ED 244 437

EC 162 450

AUTHOR

Skiba, Russell; Deno, Stanley L.

TITLE

A Correlational Analysis of the Statistical

Properties of Time-Series\_Data and Their Relationship

to Student Achievment in Resource Classrooms.

INSTITUTION

Minnesota\_Univ., Minneapolis. Inst. for Research on

Learning\_Disabilities.,

SPONS AGENCY

Special Education Programs (ED/OSERS), Washington,

REPORT NO

IRLD-RR-138

PUB DATE CONTRACT Sep 83 300-80-0622

NOTE

39p.

PUB TYPE

Reports - Research/Technical (143)

EDRS PRICE DESCRIPTORS MF01/PC02 Plus Postage.

Elementary Education; \*Learning Disabilities; \*Measurement Techniques; Research Methodology;

Statistical Distributions; Student Evaluation

#### ABSTRACT

Resource room students (N=68) in grades 1-7 of three rural and suburban Minnesota school districts participated in a study of time series data generated by a curriculum-based measurement system. A principal components factor analysis was performed to summarize relationships among the time-series properties and properties of the measurement system. In addition, multiple regression analyses were used to identify the relationship of such variables to achievement. Results indicated that the statistical characteristics of time-series data are not necessarily independent in naturally occurring data, and can be used in predicting achievement. Findings had implications in training practitioners in the use and interpretation of measurement systems based on time-series data. (Author/CL)

Reproductions supplied by EDRS are the best that can be made from the original document.



### **University of Minnesota**

Research Report No. 138

A CORRELATIONAL ANALYSIS OF THE STATISTICAL PROPERTIES OF TIME-SERIES DATA AND THEIR RELATIONSHIP TO STUDENT

ACHIEVEMENT IN RESOURCE CLASSROOMS

Russell Skiba and Stanley L. Deno

U.S. DEPARTMENT OF EDUCATION
NATIONAL INSTITUTE OF EDUCATION
EDUCATIONAL RESOURCES INFORMATION

CENTER (ERIC)
This document has been reproduced as received from the person or organization originating it

Minor changes have been made to improve reproduction quality

 Points of view or opinions stated in this document do not necessarily represent official NIE position or policy



# Institute for Research on Learning Disabilities

## IRLD

Director: James E. Ysseldyke

The Institute for Research on Learning Disabilities is supported by a contract (300-80-0622) with the Office of Special Education, Department of Education, through Title VI-G of Public Law 91-230. Institute investigators are conducting research on the assessment/decision-making/intervention process as it relates to learning disabled students.

During 1980-1983, Institute research focuses on four major areas:

- Referral
- Identification/Classification
- Intervention Planning and Progress Evaluation
- Outcome Evaluation

Additional information on the Institute's research objectives and activities may be obtained by writing to the Editor at the Institute (see Publications list for address):

The research reported herein was conducted under government sponsorship. Contractors are encouraged to express freely their professional judgment in the conduct of the project. Points of view or opinions stated do not, therefore, necessarily represent the official position of the Office of Special Education.

Research Report No. 138

# A CORRELATIONAL ANALYSIS OF THE STATISTICAL PROPERTIES OF TIME-SERIES DATA AND THEIR RELATIONSHIP TO STUDENT ACHIEVEMENT IN RESOURCE CLASSROOMS

Russell Skiba and Stanley L. Deno

Institute for Research on Learning Disabilities

University of Minnesota

September, 1983

#### Abstract

Research has shown that statistical properties of time series data can influence visual inference based on those data; little has investigated the interactions of research, however, properties in experimentally generated data. In the current study, time series data for 68 students generated by a curriculum-based measurement system were analyzed. A principal components factor analysis was performed to summarize relationships among the timeseries properties and properties of the measurement system. 'In addition, multiple regression analyses were used to identify the relationship of such variables to achievement. Results indicated that statistical characteristics of time-series data are necessarily independent in naturally occurring data, and can be used in predicting achievement. Implications for training practitioners in visual inference are discussed.

## A Correlational Analysis of the Statistical Properties of Time-Series Data and Their Relationship to Student Achievement in Resource Classrooms

In recent year, considerable controversy has been generated regarding the appropriate method for analyzing time-series data. Visual analysis of the data has been favored by many practitioners (Kratochwill, 1978). Those who favor visual inference argue that it increases the probability that only large, practically significant changes will be detected (Parsonson & Baer, 1978). Advocates of solely visual analysis also have asserted that reliance on post-hoc statistical procedures may lead researchers to ignore the necessity of maintaining strict experimental control (Michael, 1974), and may reduce the generalizability and replicability of experiments.

On the other hand, it has been argued that the statistical characteristics of time-series data make them too complex for reliable visual analysis. Jones, Weinrott, and Vaught (1978) have shown that small changes that are statistically significant may not be detected by visual inspection of the graphed data. These nondramatic but reliable effects may, in fact, prove clinically significant under some conditions (Kazdin, 1976). Finally, unique properties of time-series data, such as serial dependency (the tendency for successive data points to be correlated and thus nonindependent), may cause decreases in the reliability of visual analysis. Jones et al. (1978) found that under conditions of serial dependency, inter-observer agreement using visual analysis was only .39.

In addition to serial dependency, other properties of time-series data may affect the nature of the conclusions drawn from such data.

Jones, Vaught, and Weinrott (1977) noted that judgments made about the impact of an intervention on time-series data could be influenced by stability of baseline behavioral scores, variability of the data within and across phases, number of data points, and changes in level of performance. Kazdin (1976) identified the overall trend or slope of the data, and changes in that trend, as another important consideration.

Furlong and Wampold (1982). investigated the effects of various statistical properties of time-series data on the judgments made about those data. They generated graphs that varied along the dimensions of level, trend, scaling, and variation. Ten reviewers, members of the Journal of Applied Behavior Analysis editorial board, were instructed to sort the graphs into as many categories as they wished. Analyzing the reviewers' sorting strategies, the investigators concluded that the majority of experts classified the time-series data according to common intervention patterns. Only a minority appeared to take variability into account when making visual inferences. Further, reviewers who attended only to the absolute size of intervention effects classified the effect patterns less adequately.

In a similar study, DeProspero and Cohen (1979) generated time-series of the "ABAB reversal" type, and sent them to 250 reviewers of behavioral journals for judgments on degree of experimental control expressed by the data. Visual inspection of the data resulted in a mean inter-rater agreement of .61 between pairs of judges assigned the same graph. A mean shift in the data pattern consistent with the hypothesized effect of the experimental variables was the most

important influence on visual inference. Variability was attended to less well. Noting the wide range of opinions concerning the important characteristics of any given data display, DeProspero and Cohen concluded that a behavioral researcher relying on visual analysis "would not be likely to get the same answer twice" (p. 578) and recommended supplementing any visual analysis with statistical analysis.

Tindal, Deno, and Ysseldyke (1983) investigated the effects of various combinations of slope and variability on reliability of visual judgments, using classroom teachers as subjects. The teachers were trained in techniques of visual inference, then asked to determine the effectiveness of the program depicted on time series varying along the dimensions of slope and variability. The results indicated that slope and variability influenced visual judgments, both singly and in interaction. Reliability of visual inference was influenced by variability, and decreased most when variability increased between phases. In addition, the use of aimlines as an aid in visual analysis did not seem to have an effect on the reliability of visual analysis.

Research thus seems to indicate that statistical parameters do influence both the types, and the reliability, of judgments made through visual analysis of time-series data. Decisions made on the basis of visual inference seem most influenced by change in the mean level of performance and trend, while variability is attended to less well. Yet variability seems to have a major impact on the reliability of visual analysis.

While research has begun to investigate some of the statistical parameters that influence time-series analysis, many areas remain

**7**-

unexplored. All of the research to date has used artificially generated time series, randomly varying the data characteristics under investigation. Yet, it is possible that the statistical properties of time-series data, such as slope, variability, and trend, covary in a nonrandom fashion in data generated in experimental, clinical, or classroom situations. If this is indeed the case, it may be important to train teachers, and others who will make decisions based on the data, to attend to these characteristics and their interaction. In addition, little attention has been given to possible relationships between such characteristics and long-term outcomes, such as student achievement. It may well be that variability or number of data points are as important in predicting outcome as are more attended to attributes, such as trend or level.

The purpose of the current study was to investigate the statistical properties of time-series data generated by a curriculum based measurement system in reading. The use of actual student data allowed exploration of the relationships among various characteristics of the data. In addition, the current study was designed to determine whether properties of a system based on time-series data analysis can be used in predicting reading achievement.

#### Method

#### Subjects

The subjects were 68 resource room students in three rural and suburban Minnesota school districts. All subjects were participants in research on the effects of teachers using frequent curriculum-based measures of student performance. Subjects ranged in grade placement

from first to seventh grade; the distribution of students by grade level is shown in Table 1.

Insert Table 1 about here

All students were receiving some resource room instruction and had been receiving such special instruction for anywhere from a few months to six years ( $\bar{X}=1.96$  years). The time spent in reading instruction in the resource room ranged from 15 minutes to 105 minutes per day, with a mean of 46 minutes per day. The students' teachers averaged two years teaching experience in regular education, and five years in special education.

#### Measures

The independent variable for all analyses, student performance on a curriculum based time-series measurement system, was measured through the statistical properties of that system, described in the Procedures section. The dependent measures were two measures of achievement: timed reading samples from three third grade passages (Deno, Mirkin, & Chiang, 1982) and four subtests of the Stanford Diagnostic Reading Test (SDRT).

Achievement measures. At three different points in time during the study, three one-minute oral reading measures, consisting of randomly selected passages from the third grade level in Ginn 720, were administered to the studer. These measures were selected because of their technical adequacy (Deno et\_al., 1982) and sensitivity to change (Marston, Lowry, Deno, & Mirkin, 1981). These

curriculum-based measures had been found to be as reliable and valid as traditional standardized tests; yet more likely to reflect small changes in performance. The measurements were conducted by directing the student to begin reading at the top of the page and continue reading for one minute, at which time the examiner would say stop. If the student came to a word he/she did not know, the examiner would supply the word and prompt the reader to continue. While the student was reading, the examiner followed along on a copy of the passage and marked errors of substitution and omission. Following the reading, the numbers of words read correct and incorrect were counted and recorded, with no feedback given to the student. These three reading measures were given at the beginning of the study (pretest), in the middle, and immediately at the end of the study.

Two subtests from the <u>Stanford Diagnostic Reading Test</u> (Karlsen, Madden, & Gardner, 1976) also were given as posttest measures. The Structural Analysis and Reading Comprehension subtests were administered along with the reading passage measures. Each of the SDRT subtests has two parts, with Structural Analysis focusing on syllabication (blending and division) and Reading Comprehension focusing on answering both literal and inferential questions for previously read passages.

#### Procedures

The resource room teachers were trained in the use of the measurement procedures during a series of three half-day workshops at the beginning of the school year. Training was based on the manual, Procedures to Develop and Monitor Progress on IEP Goals (Mirkin, Deno,



Fuchs, -Wesson, Tindal, Marston, & Kuehnle, 1981). The teachers continued to use the measures over the entire school year. Visits by observers in December, February, and May, and frequent phone contacts, provided feedback to the teachers on the accuracy of their implementation of the measures.

Measurement consisted of one-minute timed samples of reading from the student's curriculum. Based on the results of previous research, the placement level for testing was set at a criteria of 20-29 words correct per minute for grades 1 and 2, and 30-39 words correct per minute for grades 3-7. Once this level was determined, passages were chosen randomly from the placement level textbook for measurement purposes. Measurements were conducted three to five times each week. Both number of words read correctly and number of errors in one minute were recorded, and plotted on an equal interval chart. Continuous graphed results allowed teachers to develop a visual record of student progress, similar to the one represented in Figure 1.

Insert Figure 1 about here

Teachers were instructed to write IEP long-range goals (LRG) using both the entry level criteria and a desired year-end mastery criteria, usually 70 words correct per minute with no more than 7 errors. The formula used in writing the long-range goal is shown in Figure 2.

#### Insert Figure 2 about here

Short-term objectives were based on the long-range goals. In order to compute the short-term objective, teachers first subtracted the baseline level of performance from the criterion level listed in the LRG. Dividing this difference by the number of weeks until the annual review, they arrived at the number of words per week gain necessary to meet the long-range goal criteria. The format used for writing short-term objectives is given in Figure 3.

Insert Figure 3 about here

In addition, the teachers were trained at the beginning of the year, and again at mid-year, in the use of the measurement procedures for evaluation of the instructional program. In order to monitor student growth, the baseline reading level and the long-range goal were connected by an aimline that showed the tudents' desired progress. Every seven data points, the teachers were to evaluate student growth using a decision rule that required use of the quarter-intersect method (White & Haring, 1980) to determine slope. An example is given in Figure 1. If the student was progressing at a rate equivalent to or greater than that indicated by the aimline, the instructional program was continued; if the projected rate of growth was less than that indicated by the aimline, the teacher was to make a substantial change in the student's program.

Student performance data on the direct repeated measures were collected and charted over a six-month period for the 68 resource room students. Variables generated from these graphed performance data for experimental analyses are discussed below.

The analyses used in the current study were correlational.

Descriptive analyses for all variables except steepness of the aimline are presented elsewhere (Skiba, Marston, Wesson, Sevcik, & Deno, 1983).

To explore the relationships among the properties of the data, a factor analysis was performed in order to summarize a large number of correlations. Variables included in the analysis were those variables available to the teacher in a visual analysis of student time-series The first subset of variables generated were those commonly in time series literature as influential in visual identified level of performance, trend, variability, and number of inference: data points. Two measures of level of performance were included: the Y-intercept (Y-INT) as an estimate of baseline performance, and the mean level of performance for the year (MY). Based on the six months of graphed performance data, the overall trend of the data was estimated by computing a regression slope for each individual (SEOPE). In order to test the probability that the slope represented a significant trend in the student's reading performance over time, and not an artifact of random variability, the c-statistic was applied to each individual time-series, as recommended by Tryon (1982). score of the c-statistic (Z-SCORE)2 was used as an estimate of the error of estimate (SEE) of the regression slope. Number of data points (DATAPTS) was measured in terms of number of measurements per week, since such a measure proved more comparable across teachers than total number of data points (which was highly influenced by beginning and end of year school schedules).

In addition, a number of variables specific to the curriculum based measurement system employed were included in the analysis, since these could be expected to influence the time series data pattern. The aimline (AIMLINE) was calculated using the formula for the shortterm objective, and served as an estimate of student progress necessary to reach the long-range goal from the Y-intercept. Student success in meeting the goal was estimated both in terms of whether the goal was met at least once (GOAL) and by the number of times the longrange goal was exceeded in the time series (NGOAL). The number of phase changes instructional was measured Бy the number of interventions (CHANGE) implemented over the course of the school year.

In order to determine whether any property or properties of the time-series data could be useful in predicting reading achievement, a series of regression analyses was performed with the achievement measures as dependent variables. The independent variables were the variables listed above: mean for the year, Y-intercept, number of data points per week, slope, number of instructional changes, standard error of estimate, z-score of c-statistic, whether the goals was met, number of times the goal was met, and steepness of the aimline. Since school achievement has been shown to be correlated most highly with

entering student ability (Bloom, 1976; Borg, 1980), two methods of controlling for achievement were used. First, pretest achievement (as measured by passage data) was forced as the first independent variable for all regression analyses conducted on the scores obtained during the third data collection. Second, two gain scores were calculated: a score representing the absolute gain in words read per minute between the first and third timed passages, and the conversion of these absolute gain scores into percentage gain. Achievement was standardized by grade or control for age effects (except for gain scores, which were based on raw data).

#### Results

Results are presented in two sections. First, the relationship among the variables was investigated by performing a factor analysis. Second, the relationship of these variables to student achievement was explored through regression analyses. Descriptive data for the 10 variables examined are presented in Table 2.

Insert Table 2 about here

#### Relationships Among the Variables

A principal components factor analysis of the time-series variables was performed. Factors with eigenvalues equal to or exceeding 1.00 were rotated, and variables that correlated with the factors at a level of .30 or greater were analyzed. The three factors retained accounted for 67.6% of the total variance. As indicated in Table 3, the items that correlated most strongly with Factor 1 have to



do with level of performance, such as the Y-intercept or the mean level of performance for the year. The standard error of estimate, and both goal-meeting variables, also loaded moderately on this factor, while number of instructional interventions loaded moderately negatively. Factor 1 accounted for 30.1% of the variance. Factor 2, explaining 23% of the variance, was most heavily influenced by the Z-score, a measure of the significance of the slope and the number of data points per week. The number of times the goal was met correlated moderately with this factor, while the Y-intercept and the standard error of estimate loaded moderately negatively on Factor 2. Factor 3 could be conceptualized as a aimline/slope factor and accounted for 14.4% of the total variance. The number of times the goal was met showed a moderate negative correlation with this factor.

Insert Table 3 about here

#### Relationship of Time-Series Variables to Achievement Variables

. . } .

Results of the regression of the time-series variables on achievement measures are presented in Table 4. As can be noted in the second column, the number of times students met their goal and the number of data points per week strongly predicted achievement on the majority of the achievement measures. It is interesting to note the increase in the proportion of residual variance explained by the time-series characteristics from the Time 2 passage scores to the Time 3 passage scores. Mean level of reading performance for the year and the Y-intercept level are moderate predictors of overall gain on the

passages (Passage Gain Score). Number of changes made in the students instructional program negatively predicted scores on the SDRT comprehension subtests and total score, while steepness of the aimline drawn by teachers at the beginning of the year negatively predicted performance on both final passage scores and overall gain in words read per minute.

Insert Table 4 about here

#### Discussion

Previous analyses of computer generated time series have indicated the influence of statistical properties of the data in visual inferences made regarding the data. Yet the majority of these studies have explored the properties in isolation, seldom taking into account interactions that may occur in experimentally generated data. The current analysis explored the relationships among the characteristics of a curriculum-based measurement system based on time-series analysis. In addition, the relationship of such characteristics to student achievement in naturally occurring time-series data was investigated.

The current results seem to indicate that the most distinctive and influential property of time-series data is the level of performance. The level of performance factor accounts for 30% of the variance in the principal component factor analysis; this figure is reminiscent of the 25%-40% of variance that Bloom (1976) suggested is accounted for by cognitive entry variables in any analysis of storent

achievement. In addition, heavy loadings of the mean for the year, as well as the Y-intercept (entry level), indicate that this factor remains important throughout the school year. As Bloom (1976) had noted, cognitive entry variables remain, unfortunately, the most powerful predictor of academic outcome.

The correlation of variability with the level of performance factor has interesting implications for time-series analysis. Future investigations of time-series data may need to take into account the tendency of scores at a higher level to show greater variability. Since increased variation has been shown to influence the accuracy of decisions made about time-series data, these results also may indicate that visual inference may be inherently more difficult at higher performance levels. These results are especially important in light of the common failure to take variability into account in visual analysis (Furlong & Wampold, 1982).

Number of changes in the student's instructional program correlated negatively with both the goal-meeting variables in Factor 1, and reading achievement. Since instructional interventions were intended to accelerate slopes and increase the probability of meeting the long-range goal, these negative correlations are, at first glance, somewhat disturbing. Given the small number of changes made in students' instructional programs, however, it is likely that the correlations reflect the fact that changes were made only for students with extremely flat slopes. In order to evaluate the effectiveness of interventions, further investigations will need to focus on changes in trend and variability resulting from individual interventions.

A positive correlation between the steepness of the aimline initially set by the teacher and the regression slope based on actual student data indicates that student time-series performance may in fact be influenced by teacher expectations. This correlation confirms the tendency noted by the investigators for time-series data to follow the aimline, and provides a caution against setting goals at too low a level. Yet, these data do not support the thesis that setting goals higher will always accelerate student performance (Lindsley, 1982). The moderate negative correlation between the number of times the goal was met and the aimline/slope factor indicates that setting goals at too high a level may actually decrease student success in reaching that goal. This conclusion is supported further by negative correlations between aimline and achievement measures: the steeper the aimline, the poorer the achievement.

Similarly, excess variability often is regarded negatively in time-series analysis. In the current analysis, variability showed no relationship to slope, although it bone a moderate negative relationship to the significance of the slope. Yet the standard error of estimate did not explain a significant proportion of the achievement variance in any of the regression analyses. Thus, while variability may play a part in determining the significance of trends in the data pattern, it does not appear to be a useful predictor of long-term outcomes.

The number of data points per week appears to be a strong predictor of both the z-score of the slope, and performance on measures of reading achievement. A number of explanations could be

practice; however, given that this practice could amount to a maximum of 3 to 5 minutes per week, this seems relatively unlikely. A more plausible suggestion might be that increased measurement provided increased performance feedback to both the student and teacher. This feedback may in turn promote subtle instructional changes or increased motivation for students to achieve their goals in reading.

Another strong predictor of reading achievement was the number of data points that exceeded the long-range goal. The long-range goals were, for the most part, set so that students would reach their goal with an average of 1 to 2 words per minute gain per week. These data thus provide some confirmation that this rate of reading growth predicts positive achievement outcomes, and may well be an appropriate rate of reading growth for students in special settings. These findings also argue for the predictive validity of frequent, curriculum-based measurement, since success on the daily measures predicted success on long-term achievement measures.

It is interesting to compare the results of the current regression analyses with similar analyses performed on this population using teacher effectiveness variables derived from process-product research (cf. Skiba, Sevcik, Wesson, King, & Deno, 1983). Only one of the 12 independent variables investigated in that study, Frequency of Correct Answers, predicted reading achievement at statistically significant levels, and the proportion of residual variance explained ranged from 1-3%. Both the number of data points per week, and the number of times the goal was met, explained a somewhat greater

proportion of residual variance, ranging as high as 28%. Variables such as academic engaged time, investigated in process-product research, have proven useful predictors of achievement in regular education; variables more directly related to monitoring student academic progress may prove to be more valuable in predicting outcomes in special education.

important implications These findings have in training practitioners in the use and interpretation of measurement systems based on time-series data. First, results indicate that properties of time-series data, such as level and variability are not independent of each other in naturally occurring data. Thus, it may be necessary to train practitioners to attend to interactions between time-series characteristics when making judgments based on visual inference. Second, in predicting long-term outcomes, properties such as frequency of measurement may be as important as more commonly attended to attributes, such as level and trend. Neither of these findings is particularly surprising, yet both represent important qualifications to the current methodology of time-series analysis.

#### References

- Bloom, B. S. <u>Human\_characteristics and school learning</u>. New York: McGraw-Hill, 1976.
- Borg, W. R. Time and school learning. In C. Denham & A. Lieberman, <u>Time to learn</u>. Washington, DC: National Institute of Education, 1980.
- Deno, S. L., Mirkin, P. K., & Chiang, B. Identifying valid measures of reading. Exceptional Children, 1982, 49, 36-47.
- DeProspero, A., & Cohen, S. Inconsistent visual analysis of intrasubject data. Journal of Applied Behavior Analysis, 1979, 12, 573-579.
- Furlong, M. J., & Wampold, B. E. Intervention effects and relative variation as dimensions in experts' use of visual analysis. Journal of Applied Behavior Analysis, 1982, 15, 415-421.
- Jones, R. R., Vaught, R. S., & Weinrott, M. Time series analysis in operant research. Journal of Applied Behavior Analysis, 1977, 10, 151-166.
- Jones, R. R., Weinrott, M. R., & Vaught, R. S. Effects of serial dependency on the agreement between visual and statistical inference. Journal of Applied Behavior Analysis, 1978, 11, 272-283.
- Karlsen, B., Madden, R., & Gardner, E. F. Stanford diagnostic reading test. New York: Harcourt Brace Jovanovich, 1976.
- Kazdin, A. E. Statistical analyses for single-case experimental design. In M. Hersen & D. Barlow (Eds.), Single case experimental designs: Strategies for studying behavior change.

  New York: Pergamon Press, 1976.
- Kratochwill, T. R. Foundations of time-series research. In T. R. Kratochwill (Ed.), <u>Single subject research: Strategies for evaluating change</u>. New York: Academic Press, 1978.
- Lindslev, O. R. How teachers could be rewarded for student learning. Paper presented at the Twelfth Annual Invitational Conference on Leadership in Special Education, Minneapolis, November, 1982.
- Marston, D., Lowry, L., Deno, S., & Mirkin, P. <u>An analysis of learning trends in simple measures of reading, spelling, and written expression: A longitudinal study</u> (Research Report No. 49). Minneapolis: University of Minnesota, Institute for Research on Learning Disabilities, 1981.

- Michael, J. Statistical inference for individual organism research: Mixed blessing or curse? Journal of Applied Behavior Analysis, 1974, 1, 647-653.
- Mirkin, P., Deno, SY, Fuchs, L., Wesson, C., Tindal, G., Marston, D., & Kuehnle, K. Procedures to develop and monitor progress on IEP goals? Minneapolis: University of Minnesota, Institute for Research on Learning Disabilities, 1981.
- Parsonson, B. S., & Baer, D. M. The analysis and presentation of graphic data. In T. R. Kratochwill (Ed.), Single subject research: Strategies for evaluating change. New York:

  Academic Press, 1978.
- Skiba, R., Marston, D., Wesson, C., Sevcik, B., & Deno, S. E.

  Characteristics of the time-series data collected through
  curriculum-based reading measurement (Research Report No. 125).

  Minneapolis; University of Minnesota, Institute for Research
  on Learning Disabilities, 1983.
- Skiba, R., Sevcik, B., Wesson, C., King, R., & Deno, S. The non-effect of process-product variables in resource classrooms (Research Report No. 121). Minneapolis: University of Minnesota, Institute for Research on Learning Disabilities, 1983.
- Tindal, G., Deno, S. L., & Ysseldyke, J. E. <u>Visual analysis of time-series data</u>: <u>Factors of influence and level of reliability</u> (Research Report No. 112). Minneapolis: University of Minnesota, Institute for Research on Learning Disabilities, 1983.
- Tryon, W. W. A simplified time-series analysis for evaluating treatment interventions. Journal of Applied Behavior Analysis, 1982, 15, 423-429.
- White, O. R., & Haring, E. G. Exceptional teaching (2nd ed.): Columbus, OH: Charles E. Merrill, 1980.



#### Footnotes

The authors gratefully acknowledge the contributions of Dr. Gerald Tindal in directing our attention to the c-statistic as a method of time-series analysis.

Tryon (1982), are:  $N_{-1}$ 

$$C = 1 - \frac{\sum_{i=1}^{N-1} (\bar{x}_i - \bar{x}_{i-1})^2}{\sum_{i=1}^{N} (\bar{x}_i - \bar{x}_i)^2}$$

where the numerator of the right hand term is the sum of the (N-1) squared consecutive differences associated with the time series. The denominator is twice the sum of the (N) squared deviations of the time-series data points from their norm.

The standard error of the c-statistic is

$$Sc = \sqrt{\frac{N+2}{(N-1)(N+1)}}$$

The c-statistic may be converted to a z-statistic and tested for significance through the following ratio:

$$Z = \frac{\bar{C}}{SC}$$



Table 1 Table

		· <del></del>	<u> </u>
Grade		Number of Students	Percentage ¿
. 1 .	Ī	2	2.9
2		18	26.5
3		15	22.1
4 -	,	16	23.5
<u>5</u> ·	Ī	, 12	17.6
6		<b>1</b> 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1	4-4
7	/	2	2.9
		1	
•	į		



Table 2 Descriptive Statistics for Time-Series Variables

	<u> </u>		<del></del>
	Mean	SD	
γ_intercept	37.01*	11.25	
Mean level for the year	51.03*	11.67	· · · · · ·
Standard error of estimate	10.04*	4.28	
Slope	1.56**	.87	
Aimline	1.68**	.53	· · · · · · · · · · · ·
Number of data points/week	2.59	. 67	
Number of instructional interventions for the year	. 67	1:14	
Z-score of the C-statistic	3.32	2.43	
Number of times goal level was exceeded	8.54	9.26	
	Frequency	Percentage	
Students reaching long-range goal at least once	58	85.3%	•

<sup>\*</sup>Expressed in words per minute \*\*Expressed in words per minute gained per week

Täble 3 Results of Principal Components Factor Analysis

		<del></del>			
Vāriāblēs <sup>ā</sup>	ī	Factor 1 <sup>h</sup>	Factor 2	Factor 3	
Y= INT		.6854	4093	0760	
MY /		_8802	0689	1592	
SEE		.5220	- 4026	-:1259	
SLOPE	e We	.2778	3362	4633	
AIMLINE		-:0053	1265	<u>.7565</u>	
DATA PTS		0087	.7380	-:1001	
CHANGE		4463	- : 0731	-:1553	
Z-SCORE		1284	. 9324	.0539	
GOAL		<u>.5460</u>	.0242	0192	
NGOAL	·	<u>.6215</u>	.4444	<u>3235</u>	

 $a_{Y-INT} = Y-intercept$ 

My = Mean level of performance for the year

SEE = Standard error of estimate of the regression slope

SLOPE = Regression slope

AIMLINE = The trend of the aimline drawn by the teacher between the Y-intercept and the long-range goal level

DATAPTS = Number of data points per week

CHANGE = Number of instructional interventions over the course of the school year

ZSCORE = The z-score of the c-statistic (see Footnote 1)
GOAL = Binary variable indicating whether the goal was met at least

NGOAL = The number of times the long-range goal level was exceeded in the time-series

byariables loading positively or negatively on the factor are underlined. The level chosen for analysis was + .30.



Table 4 Summary of Regression of Time-Series Characteristics on Achievement Measures

•	Proportion of Variance Accounted for by Pretest Achievement	Independent Variables with Significant Beta-weights	; Sign	Proportion- of Residual Variance	Proportion of Residual Variance Accounted for by All Independent Variable
Passage Score - Time 3 (N=56)	.32	NGOAL*** AIMLINE**	( <del>+)</del> (-)	.1 <u>8</u> .04	.29
SDRT - Comprehension. Subtests (N=57)	.20	CHANGE** GOAL*	(-) (+)	. 09 . 04	.21
SDRT - Structural Analys Subtests (N=57)	is ,35	·	<del></del> .	==	.05
SDRT - Total Score	.35	CHANGE**	(-)	.05	.08
Passage-Gaiā Scoreb (N=55)		NGOAL*** /-INT** DATAPTS* AIMLINE* MEAN/YEAR*	(+) (+) (+) (-) (+)	.28 .11 .04 .04 .04	. 56
Percent Gain (N=55)	i e	DATAPTS***	(+)	\23 	.35

<sup>&</sup>lt;sup>a</sup>The passage score at time I was used to control for entering achievement.

Gain in words read per minute from the third grade passages from time 1 (October) to time 3 (May).

CDATAPTS = Number of data points per week

NGOAL = Number of times goal was met CHANGE = Number of changes made in the instructional program

MEAN/YEAR= Mean level of performance for the year

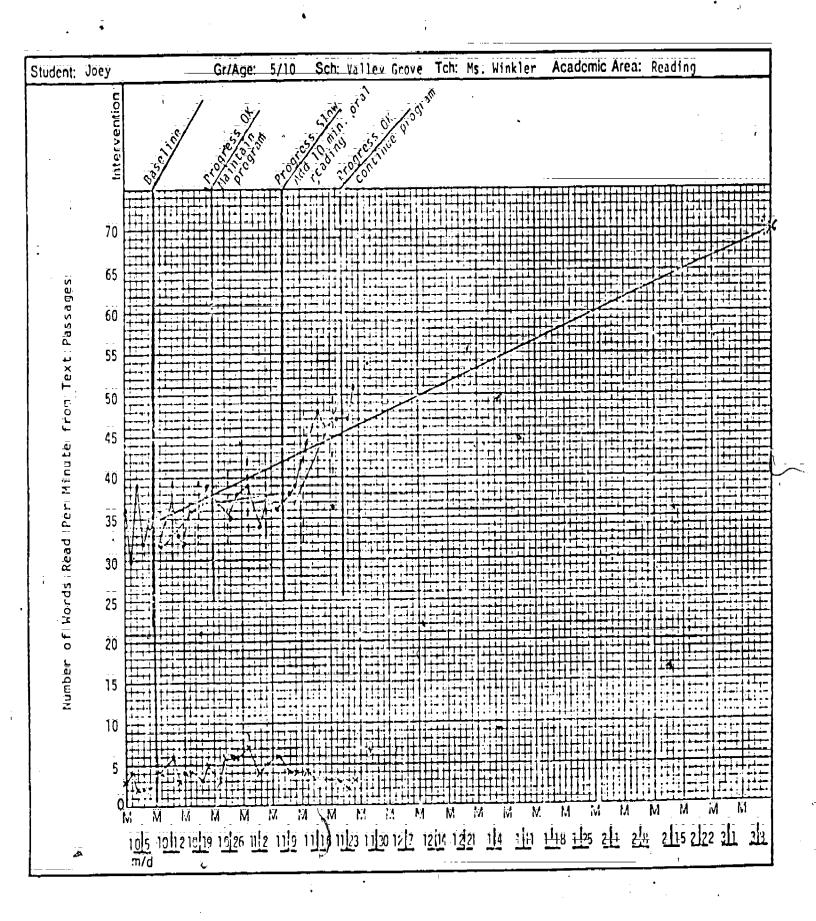
AIMLINE = Steepness, in words per week, of the aimline drawn by the teacher

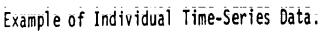
Y-INT = Y-intercept.

The proportion of the residual variance accounted for by the variable listed in Column 2, where residual variance refers to the variance remaining in the post-achievement measure after entering achievement at time 1.

<sup>&</sup>lt;sup>e</sup>The proportion of the residual variance accounted for by all independent variables after accounting for pre-achievement.

<sup>\*</sup>p < .05. \*\*p < .05. \*\*p < .05.





Condition

Behavior

Criteria

In weeks, when twill at the rate of 50 wpm or better read aloud wpm or better 5 or fewer errors.

(#) (reading series),

Figure 2. Format for Long-Range Goal: Reading

Condition	Behavior	Criteria
Each successive week, when presented with a random selection from Level (#)  of (reading series),	student will read aloud	at an average increase of  (repeated-actual performance/total # weeks) words correct/ minute and no increase in errors:

Figure 3. Format for Short-Term Objective: Reading

#### PUBLICATIONS.

Institute for Research on Learning Disabilities
University of Minnesota

The Institute is not funded for the distribution of its publications. Publications may be obtained for \$4.00 each, a fee designed to cover printing and postage costs. Only checks and money orders payable to the University of Minnesota can be accepted. All orders must be prepaid. Requests should be directed to: Editor, IRLD, 350 Elliott Hall; 75 East River Road, University of Minnesota, Minneapolis, MN 55455.

The publications listed here are only those that have been prepared since 1982. For a complete, annotated list of all IRLD publications, write to the Editor.

- Wesson, C., Mirkin, P., & Deno, S. <u>Teachers' use of self instructional</u> materials for learning procedures for developing and monitoring progress on IEP goals (Research Report No. 63). January, 1982.
- Fuchs, L., Wesson, C., Tindal, G., Mirkin, P., & Deno, S. Instructional changes, student performance, and teacher preferences: The effects of specific measurement and evaluation procedures (Research Report No. 64). January, 1982.
- Potter, M., & Mirkin, P. Instructional planning and implementation practices of elementary and secondary resource room teachers:

  Is there a difference? (Research Report No. 65). January, 1982.
- Thurlow, 'M. L., & Ysseldyke, J. E. Teachers' beliefs about LD students (Research Report No. 66). January, 1982.
- Graden, J., Thurlow, M. L., & Ysseldyke, J. E. Academic engaged time and its relationship to learning: A review of the literature (Monograph No. 17). January, 1982.
- King, R., Wesson, C., & Deno, S. <u>Birect and frequent measurement of student performance</u>: <u>Does it take too much time</u>? (Research Report No. 67). February, 1982.
- Greener, J. W., & Thurlow, M. L. Teacher opinions about professional education training programs (Research Report No. 68). March, 1982.
- Algozzine, B., & Ysseldyke, J. Learning disabilities as a subset of school failure: The oversophistication of a concept (Research Report No. 69). March, 1982.
- Fuchs, D., Zern, D. S., & Fuchs, L. S. A microanalysis of participant behavior in familiar and unfamiliar test conditions (Research Report No. 70). March, 1982.

- Shinn, M. R., Ysseldyke, J., Deno, S., & Tindal, G. A comparison of sychometric and functional differences between students labeled learning disabled and low achieving (Research Report No. 71).

  March, 1982.
- Thurlow, M. L. Graden, J., Greener, J. W., & Ysseldyke, J. E. Academic responding time for LD and non-LD students (Research Report No. 72). April, 1982.
- Graden, J., Thurlow, M., & Ysseldyke, J. <u>Instructional ecology and</u> academic responding time for students at three levels of teacher-perceived behavioral competence (Research Report No. 73). April, 1982.
- Algozzine, B., Ysseldyke, J., & Christenson, S. The influence of teachers' tolerances for specific kinds of behaviors on their ratings of a third grade student (Research Report No. 74).

  April, 1982.
- Wesson, C., Deno, S., & Mirkin, P. Research on developing and monitoring progress on IEP goals: Current findings and implications for practice (Monograph No. 18). April, 1982.
- Mirkin, P., Marston, D., & Deno, S. L. Direct and repeated measurement of academic skills: An alternative to traditional screening, referral, and identification of learning disabled students (Research Report No. 75). May, 1982.
- Algozzine, B., Ysseldyke, J., Christenson, S., & Thurlow, M. <u>Teachers'</u> intervention choices for children exhibiting different behaviors in school (Research Report No. 76). June, 1982.
- Tucker, J., Stevens, L. J., & Ysseldyke, J. E. <u>Learning disabilities:</u>
  The experts speak out (Research Report No. 77). June, 1982.
- Thurlow, M. L., Ysseldyke, J. E., Graden, J., Greener, J. W., & Mecklenberg, C. Mcademic responding time for LD students receiving different levels of special education services (Research Report No. 78). June, 1982.
- Graden, J. L., Thurlow, M. L., Ysseldyke, J. E., & Algozzine, B. <u>Instructional ecology and academic responding time for students in different reading groups</u> (Research Report No. 79). July, 1982.
- Mirkin, P. K., & Potter, M. L. A survey of program planning and implementation practices of LD teachers (Research Report No. 80). July, 1982.
- Fuchs, L. S., Fuchs, D., & Warren, L. M. Special education practice in evaluating student progress toward goals (Research Report No. 81). July, 1982.
- Kuehnle, K., Deno, S. L., & Mirkin, P. K. Behavioral measurement of social adjustment: What behaviors? What setting? (Research Report No. 82). July, 1982.

- Fuchs, D., Dailey, Ann Madsen, & Fuchs, L. S. Examiner familiarity and the relation between qualitative and quantitative indices of expressive language (Research Report No. 83). July, 1982.
- Videen, J., Deno, S., & Marston, D. Correct word sequences: A valid indicator of proficiency in written expression (Research Report No. 84). July, 1982.
- Potter, M. L. Application of a decision theory model to eligibility and classification decisions in special education (Research Report No. 85). July, 1982.
- Greener, J. E., Thurlow, M. L., Graden, J. L., & Ysseldyke, J. E. The educational environment and students' responding times as a function of students' teacher-perceived academic competence (Research Report No. 86). August, 1982.
- Deno, S., Marston, D., Mirkin, P., Lowry, L., Sindelar, P., & Jenkins, J.

  The use of standard tasks to measure achievement in reading, spelling, and written expression: A normative and developmental study (Research Report No. 87). August, 1982.
- Skiba, R., Wesson, C., & Deno, S. L. The effects of training teachers in the use of formative evaluation in reading: An experimental-control comparison (Research Report No. 88). September, 1982.
- Marston, D., Tindal, G., & Deno, S. L. <u>Eligibility for learning disability services</u>: A direct and repeated measurement approach (Research Report No. 89). September, 1982.
- Thurlow, M. L., Ysseldyke, J. E., & Graden, J. L. LD students' active academic responding in regular and resource classrooms (Research Report No. 90). September, 1982.
- Ysseldyke, J. E., Christenson, S., Pianta, R., Thurlow, M. L., & Algozzine, B. An analysis of current practice in referring students for psycho-educational evaluation: Implications for change (Research Report No. 91). October, 1982.
- Ysseldyke, J. E., Algozzine, B., & Epps, S. A logical and empirical analysis of current practices in classifying students as handicapped (Research Report No. 92). October, 1982.
- Tindal, G., Marston, D., Deno, S. L., & Germann, G. Curriculum differences in direct repeated measures of reading (Research Report No. 93). October, 1982.
- the reliability of simple direct measures of academic performance (Research Report No. 94). October, 1982.
- Ysseldyke, J. E., Thurlow, M. L., Mecklenburg, C., & Graden, J. Observed changes in instruction and student responding as a function of referral and special education placement (Research Cort No. 95).

  October, 1982.

iit.

- Fuchs, L. S., Deno, S. L., & Mirkin, P. K. Effects of frequent curriculum-based measurement and evaluation on student achievement and knowledge of performance: An experimental study (Research Report No. 96). November, 1982.
- Fuchs, L. S., Deno, S. L., & Mirkin, P. K. Direct and frequent measurement and evaluation: Effects on instruction and estimates of student progress (Research Report No. 97). November, 1982.
- Tindal, G., Wesson, C., Germann, G., Deno, S. L., & Mirkin, P. K. The Pine County model for special education delivery: A data-based system (Monograph No. 19). November, 1982.
- Epps, S., Ysseldyke, J. E., & Algozzine, B. An analysis of the conceptual framework underlying definitions of learning disabilities (Research Report No. 98). November, 1982.
- Epps, S., Ysseldyke, J. E., & Algozzine, B. <u>Public-policy implications</u>
  of different definitions of learning disabilities (Research Report No. 99). November, 1982.
- Ysseldyke, J. E., Thurlow, M. L., Graden, J. L., Wesson, C., Deno, S. L., & Algozzine, B. Generalizations from five years of research on assessment and decision making (Research Report No. 100). November, 1982.
- Marston, D., & Deno, S. L. Measuring academic progress of students with learning difficulties: A comparison of the semi-logarithmic chart and equal interval graph paper (Research Report No. 101). November, 1982.
- Beattie, S., Grise, P., & Algozzine, B. Effects of test modifications on minimum competency test performance of third grade learning disabled students (Research Report No. 102). December, 1982
- Algozzine, B., Ysseldyke, J. E., & Christenson, S. An analysis of the incidence of special class placement: The masses are burgeoning (Research Report No. 103). December, 1982.
- Marston, D., Tindal, G., & Deno, S. L. Predictive efficiency of direct, repeated measurement: An analysis of cost and accuracy in classification (Research Report No. 104). December, 1982.
- Wesson, C., Deno, S., Mirkin, P., Sevcik, B., Skiba, R., King, R.,

  Tindal, G., & Maruyama, G. Teaching structure and student achievement effects of curriculum-based measurement: A causal (structural)
  analysis (Research Report No. 105). Becember, 1982.
- Mirkin, P. K., Fuchs, L. S., & Deno, S. L. (Eds.). Considerations for designing a continuous evaluation system: An integrative review (Monograph No. 20). December, 1982.
- Marston, D., & Deno, S. L. <u>Implementation of direct and repeated</u> measurement in the school setting (Research Report No. 106).

  December, 1982.



- Deno, S. L., King, R., Skiba, R., Sevcik, B., & Wesson, C. The structure of instruction rating scale (SIRS): Development and technical characteristics (Research Report No. 107). January, 1983.
- Thurlow, M. L., Ysseldyke, J. E., & Casey, A. <u>Criteria for identifying</u>
  <u>LD students: Definitional problems exemplified</u> (Research Report
  No. 108). January, 1983.
- Tindal, G., Marston, D., & Deno, S. L. The reliability of direct and repeated measurement (Research Report No. 109). February, 1983.
- Fuchs, D., Fuchs, L. S., Dailey, A. M., & Power, M. H. Effects of pretest contact with experienced and inexperienced examiners on handicapped children's performance (Research Report No. 110). February, 1983
- King, R. P., Deno, S., Mirkin, P., & Wesson, C. The effects of training teachers in the use of formative evaluation in reading: An experimental-control comparison (Research Report No. 111). February, 1983.
- Tindal, G., Deno, S. L., & Ysseldyke, J. E. <u>Visual analysis of time</u>
  <u>series data: Factors of influence and level of reliability</u> (Research
  Report No. 112). March, 1983.
- Tindal, G, Shinn, M., Fuchs, L., Fuchs, D., Deno, S., & Germann, G. The technical adequacy of a basal reading series mastery test (Research Report No. 113). April, 1983.
- Sevcik, B., Skiba, R., Tindal, G., King, R., Wesson, C., Mirkin, P., & Deno, S. Communication of IEP goals and student progress among parents, regular classroom teachers, and administrators using systematic formative evaluation (Research Report No. 114). April, 1983.
- Wesson, C. Two student self-management techniques applied to data-based program modification (Research Report No. 115). April, 1983.
- Wesson, C., Skiba, R., Sevcik, B., King, R., Tindal, G., Mirkin, P., & Deno, S. The impact of the structure of instruction and the use of technically adequate instructional data on reading improvement (Research Report No. 116). May, 1983.
- Wesson, C. Teacher vs student selection of instructional activities (Research Report No. 117). May, 1983.
- Tindal, G., & Deno, S. Factors influencing the agreement between visual and statistical analyses of time series data (Research Report No. 118). June, 1983.
- Skiba, R. S. Classroom behavior management: A review of the literature (Monograph No. 21), June, 1983.
- Graden, J. L., Thurlow, M. L., & Ysseldyke, J. E. When are students most academically engaged? Academic responding time in different instructional ecologies (Research Report No. 119). June, 1983.

- Fuchs, L. S., Deno, S. L., & Roettger, A. The effect of alternative data-utilization rules on spelling achievement: An n of 1 study (Research Report No. 120). June, 1983.
- Skiba, R., Sevcik, B., Wesson, C., King, R., & Deno, S. The non-effect of process-product variables in resource classrooms (Research Report No. 121). June, 1983.
- Fuchs, L. Tindal, G., Shinn, M., Fuchs, D., Deno, S., & Germann, G. <u>Tech-nical</u> adequacy of basal readers' mastery tests: The Ginn 720 series (Research Report No. 122). June, 1983.
- Tindal, G., Germann, G., Marston, D., & Deno, S. The effectiveness of special education: A direct measurement approach (Research Report No. 123). June, 1983.

- Sevcik, B., Skiba, R., Tindal, G., King, R., Wesson, C., Mirkin, P., & Deno, S. Curriculum-based measurement: Effects on instruction, teacher estimates of student progress, and student knowledge of performance (Research Report No. 124). July, 1983.
- Skiba, R., Marston, D., Wesson, C., Sevcik, B., & Deno, S. L. Characteristics of the time-series data collected through curriculum-based reading measurement (Research Report No. 125). July, 1983.
- Ysseldyke, J., Christenson, S., Graden, J., & Hill, D. Practical implications of research on referral and opportunity to learn (Monograph No. 22). July, 1983.
- Marston, D., Deno, S., & Tindal, G. A comparison of standardized achievement tests and direct measurement techniques in measuring pupil progress (Research Report No. 126). July, 1983.
- Fuchs, D., Fuchs, L. S., Tindal, G., & Deno, S. L. <u>Variability of performance: A "signature" characteristic of learning disabled children?</u>
  (Research Report No. 127). July, 1983.
- Tindal, G., Fuchs, L., Fuchs, D., Shinn, M., Deno, S., & Germann, G. The technical adequacy of a basal series mastery test: The Scott-Foresman reading program (Research Report No. 128). July, 1983.
- Fuchs, L. S., Fuchs, D., & Deno, S. L. The nature of inaccuracy among readability formulas (Research Report No. 129). July, 1983.
- Fuchs, L., Tindal, G., Fuchs, D., Shinn, M., Deno, S., & Germann, G. The technical adequacy of a basal reading mastery test: The Holt basic reading series (Research Report No. 130). July, 1983.
- Ysseldyke, J. E., Christenson, S., Algozzine, B., & Thurlow, M. L. <u>Class-room teachers' attributions for students exhibiting different behaviors</u> (Research Report No. 131). July, 1983.

- Tindal, G., Germann, G., & Deno, S. <u>Descriptive research on the Pine County norms: A compilation of findings</u> (Research Report No. 132); July, 1983.
- Shiba, R. J. The relationship between classroom management strategies and student misbehaviors (Research Report No. 133). July, 1983.
- Fuchs, D., Fuchs, L. S., Power, M. H., & Dailey, A. M. Systematic bias in the assessment of handicapped children (Research Report No. 134).

  July, 1983.
- Fuchs, D., & Fuchs, L. S. The importance of scorer bias to handicapped preschoolers'-stronger performance with familiar examiners (Research Report No. 135). July, 1983.
- Foster, G. G., Ysseldyke, J. E., Casey, A., & Thurlow, M. L. The congruence between reason for referral and placement outcome (Research Report No. 136). August, 1983.
- Potter, M. L. Instructional decision-making practices of teachers of learning disabled students (Research Report No. 137). September, 1983.
- Skiba, R., & Deno, S. A correlational analysis of the statistical properties of time-series data and their relationship to student achievement in resource classrooms (Research Report No. 138). September, 1983.
- Potter, M. L. <u>Decision research and its application to educational</u> settings: A <u>literature review</u> (Monograph No. 23). September, 1983.



